

PSS Curs 13

D.S.P. ($\hat{w}(z)$)

Densitate spectr. de putere

$x[n]$

$\hat{\Gamma}_{xx}[m]$

$\Gamma_{xx}(z)$

$\xleftrightarrow{z=e^{j\omega}}$

$\Gamma_{xx}(\omega)$

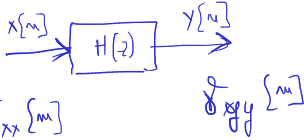
Exemplu

zg. alb
 $w[n]$

$\hat{\Gamma}_{ww}[m] = \delta[m]$

$\Gamma_{ww}(z) = \sigma_w^2 = c$

$\Gamma_{ww}(\omega) = \sigma_w^2 = c$



$\hat{\Gamma}_{xx}[m]$

$\hat{\Gamma}_{yy}[m]$

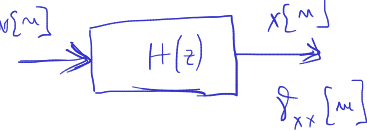
$$\Gamma_{xx}(z) \cdot H(z) \cdot H(z^*) = \Gamma_{yy}(z)$$

$$\Gamma_{xx}(z) = \sigma_w^2 \cdot H(z) \cdot H^*\left(\frac{1}{z^*}\right)$$

(=)

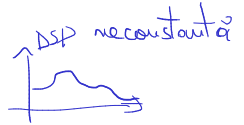
zg. alb

$w[n]$



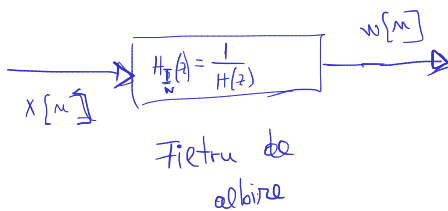
Albator, me-alb

$$\sigma_w^2 \cdot H(z) \cdot H(z^*) = \Gamma_{xx}(z)$$

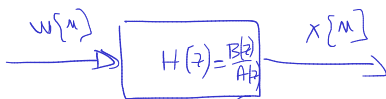


$H(z)$ = stabil (și pole și zerourile în arcul unitate)

$$\Gamma_{xx}(z) = 3 \cdot \frac{(z-0.5)(z-2)(z+0.75)(z+\frac{1}{0.75})}{(z-2.1)(z-\frac{1}{2.1})(z-1.8)(z-\frac{1}{1.8})}$$



Filtru de albire



Ec. cu diferențe: $x[n] = -a_1 \cdot x[n-1] - a_2 \cdot x[n-2] - \dots - a_p \cdot x[n-p] + b_0 \cdot w[n] + b_1 \cdot w[n-1] + \dots + b_q \cdot w[n-q]$

$$x[n] + \sum_{k=1}^p a_k \cdot x[n-k] = \sum_{k=0}^q b_k \cdot w[n-k]$$

ARMA(z, p)

$$\boxed{\text{Autoregresiv AR}(p)} \hat{=} H(z) = \frac{b_0 z^0}{A(z)} = \frac{1}{1 + a_1 z^{-1} + \dots + a_p z^{-p}}$$

$$\text{Ec. cu dif. : } x[n] + \sum_{k=1}^p a_k x[n-k] = w[n] \Leftrightarrow \boxed{x[n] + a_1 x[n-1] + \dots + a_p x[n-p] = w[n]}$$

Leg. între param. modelului (a_k, b_k) și $\gamma_{xx}[m]$

$$x[n] = - \sum_{k=1}^p a_k x[n-k] + \sum_{k=0}^q b_k w[n-k]$$

$$E\{a+b\} = E\{a\} + E\{b\}$$

$$E\{c \cdot x\} = c \cdot E\{x\}$$

$$\gamma_{xx}[m] = \overline{x[n] \cdot x[n-m]} = E\{x[n] \cdot x[n-m]\} = E\left\{ - \sum_{k=1}^p a_k \cdot \underbrace{x[n-k] \cdot x[n-m]}_{\gamma_{xx}[m-k]} + \sum_{k=0}^q b_k \underbrace{w[n-k] \cdot x[n-m]}_{\gamma_{wx}[m-k]} \right\}$$

$$\gamma_{xx}[m] = - \sum_{k=1}^p a_k \cdot E\{x[n-k] \cdot x[n-m]\} + \sum_{k=0}^q b_k \cdot E\{w[n-k] \cdot x[n-m]\}$$

$$\gamma_{xx}[m] = - \sum_{k=1}^p a_k \cdot \gamma_{xx}[m-k] + \sum_{k=0}^q b_k \cdot \gamma_{wx}[m-k]$$

$$\gamma_{xx}[m] = - \sum_{k=1}^p a_k \gamma_{xx}[m-k] + \sum_{k=0}^q b_k \gamma_{wx}[m-k] \leftarrow \text{In general, not ARMA}$$

Pentru AR: $b_0 = 1, b_{1,2,3} = 0$

$$\gamma_{wx}[m]$$

$$\gamma_{xx}[m] = - \sum_{k=1}^p a_k \gamma_{xx}[m-k] + \begin{cases} 0, & m > q \\ \sigma_w^2 \cdot \sum_{k=0}^{q-m} h[k] b_{k+m}, & 0 \leq m \leq q \end{cases}$$

$$\gamma_{xx}[-m] = \gamma_{xx}[m]$$

Pentru procese AR:

$$\gamma_{xx}[m] = \begin{cases} - \sum_{k=1}^p a_k \gamma_{xx}[m-k], & m > 0 \\ - \sum_{k=1}^p a_k \gamma_{xx}[m-k] + \sigma_w^2, & m = 0 \\ \gamma_{xx}^*[-m], & m < 0 \end{cases}$$

$$\underline{m=0} : \gamma_{xx}[0] = - \sum_{k=1}^p a_k \gamma_{xx}[0-k] + \sigma_w^2$$

$$\begin{aligned} \underline{m=1} : & \gamma_{xx}[1] + a_1 \gamma_{xx}[0] + a_2 \gamma_{xx}[-1] + \dots + a_p \gamma_{xx}[-p+1] = 0 \\ \underline{m=2} : & \gamma_{xx}[2] + a_1 \gamma_{xx}[1] + a_2 \gamma_{xx}[0] + \dots + a_p \gamma_{xx}[-p+2] = 0 \\ & \vdots \\ \underline{m=p} : & \gamma_{xx}[p] + a_1 \gamma_{xx}[p-1] + a_2 \gamma_{xx}[p-2] + \dots + a_p \gamma_{xx}[0] = 0 \end{aligned}$$

Ecuațiile Yule-Walker
(ec. normale)

$$\begin{bmatrix} s_{xx}[0] & s_{xx}[-1] & s_{xx}[-2] & \dots & s_{xx}[-p] \\ s_{xx}[1] & s_{xx}[0] & \dots & \dots & s_{xx}[-p+1] \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ s_{xx}[p] & s_{xx}[p-1] & \dots & \dots & s_{xx}[0] \end{bmatrix} \cdot \begin{bmatrix} 1 \\ a_1 \\ a_2 \\ \vdots \\ a_p \end{bmatrix} = \begin{bmatrix} w \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$A \cdot X = B$$

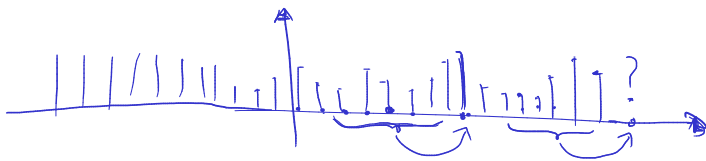
$$w[n] \xrightarrow{H(z)} x[n]$$

$$H(z) = \frac{1}{1 + a_1 z^{-1} + \dots + a_p z^{-p}}$$

Predicție liniară (f.f.f. scurt)

$$x[n] = \{ 1, 2, 3, 2, 2, 2, 3, 1, 2, 3 \} \quad ?$$

ce val. urmează?



$$x[n] \approx a_1 x[n-1] + a_2 x[n-2] + \dots + a_p x[n-p]$$

$\hat{x}[n] = x[n]$ prezis

$$x[n] = \underbrace{a_1 x[n-1] + a_2 x[n-2] + \dots + a_p x[n-p]}_{\hat{x}[n]} + e[n]$$

eroare de predicție

$$x[n] - a_1 x[n-1] - a_2 x[n-2] - \dots - a_p x[n-p] = e[n]$$

eroare de predicție
zgomot alb

Cum calculăm a_1, a_p ?

Vreau $(x[n] - \hat{x}[n])$ cât mai mică
adov. prezis

Vreau E_{eroare} $\mathcal{E} = \sum_n (x[n] - \hat{x}[n])^2$ cât mai mic \Rightarrow

Vreau ca media $\frac{(x[n] - \hat{x}[n])^2}{2}$ cât mai mică

Vrem găsi cei mai buni coef. $\{a_p\}$ din condiția : $\overline{(x[n] - \hat{x}[n])^2}$ să fie minim

$$E\{(x[n] - \hat{x}[n])^2\} \text{ vrem minim}$$

$$\Leftrightarrow \text{Vrem ca } E\left\{\underbrace{\left(x[n] - \underbrace{a_1 x[n-1] - a_2 x[n-2] - \dots - a_p x[n-p]}_A\right)^2}_{\text{minim}}\right\}$$

Vrem a_1, a_2, \dots, a_p astfel încât $A = \text{minim}$:

$$\begin{cases} \frac{\partial A}{\partial a_1} = 0 \\ \frac{\partial A}{\partial a_2} = 0 \\ \vdots \\ \frac{\partial A}{\partial a_p} = 0 \end{cases}$$

$$\frac{\partial A}{\partial a_2} = 0 \Leftrightarrow E\left\{\frac{\partial}{\partial a_2} \left(\dots \right)\right\} = 0$$

$$\frac{\partial}{\partial x} E\{f(x)\} = E\left\{\frac{\partial f}{\partial x}\right\}$$

$$E\left\{2 \cdot (x[n] - a_1 x[n-1] - \dots - a_p x[n-p]) \cdot (-x[n-1])\right\} = 0$$

$$E\left\{x[n] \cdot x[n-1] - a_1 x[n-1] x[n-1] - \dots - a_p x[n-p] x[n-1]\right\} = 0$$

$$E\{x[n] \cdot x[n-1]\} - a_1 E\{x[n-1] x[n-1]\} - \dots - a_p E\{x[n-p] x[n-1]\} = 0$$

$\delta_{xx}[1] \quad \delta_{xx}[0] \quad \delta_{xx}[-p+1]$

$$\rightarrow \delta_{xx}[1] - a_1 \delta_{xx}[0] - a_2 \delta_{xx}[-1] - \dots - a_p \delta_{xx}[-p+1] = 0$$

$$\frac{\partial A}{\partial a_2} = 0 \rightarrow \delta_{xx}[2] - a_1 \delta_{xx}[1] - a_2 \delta_{xx}[0] - \dots - a_p \delta_{xx}[-p+2] = 0$$

$$\frac{\partial A}{\partial a_p} = 0 \rightarrow$$

Ec. Yule-Walker :

$$\begin{cases} \delta_{xx}[0] + a_1 \delta_{xx}[-1] + a_2 \delta_{xx}[-2] + \dots + a_p \delta_{xx}[-p] = \sigma_w^2 \\ \delta_{xx}[1] + a_1 \delta_{xx}[0] + a_2 \delta_{xx}[-1] + \dots + a_p \delta_{xx}[-p+1] = 0 \\ \delta_{xx}[2] + a_1 \delta_{xx}[1] + a_2 \delta_{xx}[0] + \dots + a_p \delta_{xx}[-p+2] = 0 \\ \vdots \\ \delta_{xx}[p] + a_1 \delta_{xx}[p-1] + a_2 \delta_{xx}[p-2] + \dots + a_p \delta_{xx}[0] = 0 \end{cases}$$